STAT C206B: Topics in Stochastic Processes

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1 Lecture 1 2020-01-21

Ferguson distributions / Dirichlet processes

Definition 1 (Gamma distribution)

Random variable X supported on $(0, \infty)$ has Gamma distribution with shape $\alpha > 0$ and inverse scale / rate $\beta > 0$, written $X \sim \text{Gamma}(\alpha, \beta)$ if it has density

$$f_X(t) = \mathbb{1}\{t \in (0, \infty)\} \frac{\beta^{\alpha} t^{\alpha - 1} e^{-\beta t}}{\Gamma(\alpha)}$$
(1)

where $\Gamma(t) = \int_0^\infty u^{t-1} e^{-u} du$ is the Gamma function defined for all $\Re t > 0$ and analytically continued to $\mathbb{C} \setminus \{ n \in \mathbb{Z} : n < 0 \}$

Proposition 2 (Gamma closed under summation)

If $Y \sim \text{Gamma}(\alpha, \beta)$ and $Z \sim \text{Gamma}(\gamma, \beta)$ are independent, then $Y + Z \sim \Gamma(\alpha + \gamma, \beta)$.

closed-sum}

Proof.

{prop:gamma-

 $f_{Y+Z}(t) = \int_0^t f_Y(u) f_Z(t-u) du$ $= \frac{1}{\Gamma(\alpha)\Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} \int_0^t u^{\alpha-1} (t-u)^{\gamma-1} du$

$$= \frac{1}{\Gamma(\alpha)\Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} \int_{0}^{1} (tv)^{\alpha-1} (t-(tv))^{\gamma-1} t dv$$

$$= \frac{1}{\Gamma(\alpha)\Gamma(\gamma)} \beta^{\alpha+\gamma} e^{-\beta t} t^{\alpha+\gamma-1} B(\alpha, \gamma)$$

$$=\frac{1}{\Gamma(\alpha)\Gamma(\gamma)}\beta^{\alpha+\gamma}e^{-\beta t}t^{\alpha+\gamma-1}B(\alpha,\gamma)$$

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where
$$B(x,y) = \int_0^1 t^{x-1} (1-t)^{y-1} dt = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)}$$
 is the beta function

A closely related distribution obtained from concatenating Gamma random variables into a vector and then normalizing the sum to 1 is the Dirichlet distribution.

Definition 3 (Dirichlet distribution)

Let $\alpha \in (0, \infty)^K$. Random (probability) vector X taking values on the K-1-dimensional probability simplex $\Delta^{K-1} = \{ \boldsymbol{x} \in [0,1]^K : \sum_i x_i = 1 \}$ has Dirichlet distribution of order K and concentration parameters $\boldsymbol{\alpha}$, denoted $X \sim \text{Dir}(\boldsymbol{\alpha})$, if it has density

$$f_X(\boldsymbol{x}) = \mathbb{1}\{\boldsymbol{x} \in \Delta\} \underbrace{\frac{\Gamma\left(\sum_{i=1}^K \alpha_i\right)}{\prod_{i=1}^K \Gamma(\alpha_i)}}_{=:B(\boldsymbol{\alpha})^{-1}} \prod_{i=1}^K x_i^{\alpha_i - 1}$$

Proposition 4 (Constructing Dirichlet from Gammas)

Let X_1, \ldots, X_n be independent Gamma (α_i, β) distributed, $S_n = \sum_{i=1}^n X_i$. Then $(V_i)_i = (X_i/S_n)_i \sim$ {prop:dirich

let-from-gam ma}

Proof. $S_n \sim \Gamma(\sum_{i=1}^n \alpha_i, \beta)$ by Proposition 2 and for $v \in \Delta^{n-1}$, we have

$$\begin{split} f_V(\boldsymbol{v}) &= \int_0^\infty f_X\left(sv_1,\dots,sv_{n-1},sv_n\right) f_{S_n}(s) ds \\ &= \int_0^\infty e^{-\sum_{i=1}^n sv_i} \left(\prod_{i=1}^n \frac{\left(sv_i\right)^{\alpha_i-1}}{\Gamma(\alpha_i)}\right) \frac{s^{\sum_i^n \alpha_i-1} e^{-s}}{\Gamma(\sum_i^n \alpha_i)} ds \\ &= \frac{1}{\prod_1^n \Gamma(\alpha_i)} \prod_{i=1}^n v_i^{\alpha_i-1} \int_0^\infty e^{-s\sum_1^n \sigma_i^* \int_S^1 \left(\sum_1^n \alpha_i\right) - 1} ds \\ &= \frac{\Gamma(\sum_{i=1}^n \alpha_i)}{\prod_1^n \Gamma(\alpha_i)} \prod_{i=1}^n v_i^{\alpha_i-1} \end{split}$$

Similar to Proposition 2 (Gamma closed under summation), where adding two Gammas yielded another Gamma where the parameters were added, Dirichlet distributions enjoy a similar kind of closure: "clumping" coordinate axes together (described below) yields another Dirichlet distribution where the parameters of the clumped axes are summed together.

Proposition 5 (Dirichlet clumping property)

Suppose $X \sim \text{Dir}(\alpha_1, \dots, \alpha_n)$. For any $r \leq n$, let $V_i = X_i$ for $i \in [r]$ and let $V_{r+1} = \sum_{j=r+1}^n X_j$. Then $V \sim \operatorname{Dir}(\alpha_1, \dots, \alpha_r, \sum_{j=r+1}^n \alpha_j).$

Proof. By induction, it suffices to show this for r = n - 2. Notice

$$f(v_1, \dots, v_r, s) = B(\alpha)^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) \int \mathbb{1} \left\{ x_{n-1} + x_n = s \right\} x_{n-1}^{\alpha_{n-1} - 1} x_n^{\alpha_n - 1} dx_{n-1} dx_n$$

$$= B(\alpha)^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) \int_0^s u^{\alpha_{n-1} - 1} (s - u)^{\alpha_n - 1} du$$

$$= B(\alpha)^{-1} \left(\prod_{i=1}^{n-1} v_i^{\alpha_i - 1} \right) s^{\alpha_{n-1} + \alpha_n - 1} B(\alpha_{n-1}, \alpha_n)$$

Since $\frac{B(\alpha_{n-1},\alpha_n)}{B(\alpha)} = \frac{\Gamma(\sum_{i=1}^{n} \alpha_i)}{\Gamma(\alpha_{n-1}+\alpha_n)\prod_{i=1}^{n-2} \Gamma(\alpha_i)}$, we are done.

Iterating this result over coordinate axes other than the last n-r, we see that "clumping together" entries in a Dirichlet random vector yields another Dirichlet random vector with parameters also "clumped together." Concretely, for any mapping $\phi: [n+1] \to [m+1]$ if $U_j = \sum_{\phi(i)=j} V_i$ then U has Dirichlet distribution with parameters $\gamma_j = \sum_{\phi(i)=j} \alpha_i$.

Generalizing this clumping property is the motivation for $Ferguson\ Distributions$ [Fer73].

Definition 6 (Ferguson / Dirichlet process distribution)

Let μ be a finite positive Borel measure on complete separable metric space E. A random probability measure μ^* on E (i.e. a stochastic process indexed by a σ -algebra on E) has Ferguson distribution with parameter μ if for every finite partition $(B_i)_{i \in [r]}$ of E the random vector

$$(\mu^*(B_i))_{i\in[r]} \sim \operatorname{Dir}(\mu(B_1),\ldots,\mu(B_r))$$

Lemma 7 (Preservation of Ferguson under pushforward)

Let μ^* be Ferguson with parameter μ and $\phi: E \to F$ measurable. Then the pushforward $\mu^* \circ \phi^{-1}$ is a random probability measure on F that has Ferguson distribution with parameter $\mu \circ \phi^{-1}$.

Proof. For $(B_i)_{i\in[r]}$ a finite partition of F, $(\phi^{-1}(B_i))_i$ is a finite partition of E. Since μ^* is Ferguson

$$(\mu^*(\phi^{-1}(B_i)))_i \sim \text{Dir}((\mu(\phi^{-1}(B_i)))_i)$$

Hence $\mu^* \circ \phi^{-1}$ is Ferguson with parameter $\mu \circ \phi^{-1}$.

1.1.1 Connection to Polya urns

Next, we turn to an important relation between Ferguson distributions and a generalized Pólya urn. Our presentation draws from material originally in [BM73].

Definition 8 (Polya sequence)

A sequence $(X_n)_{n\in\mathbb{N}}$ with values in E is a Polya sequence with parameter μ if for all $B\subset E$.

$$\Pr[X_1 \in B] = \mu(B)/\mu(E)$$

$$\Pr[X_{n+1} \in B \mid X_1, \dots, X_n] = \mu_n(B)/\mu_n(E)$$

where $\mu_n = \mu + \sum_{i=1}^n \delta_{X_i}$.

Remark 9. When E is finite (e.g. a set of colors for the balls), (X_n) represents the result of successive draws from an urn with initially $\mu(x)$ balls of color $x \in E$ and after each draw a ball of the same color as the one drawn is added back to give an urn with color distribution $\mu_{n+1}(x)$.

 $\begin{tabular}{l} blackwell11973 \\ \hline [BM73] gives the following result connecting Pólya sequences and Ferguson distributions. \end{tabular}$

Theorem 10 (Polya Urn Schemes)

Let (X_n) be a Polya sequence with parameter μ . Then:

- 1. $m_n = \mu_n/\mu_n(E)$ converges almost surely to a limiting discrete measure μ^*
- 2. μ^* has Ferguson distribution with parameter μ
- 3. Given μ^* , $(X_i)_{i>1}$ are independent with distribution μ^*

Proof. First consider E finite and let μ^* and $\{X_i\}$ be random variables whose joint distribution satisfies (2.) and (3.). Let π_n be empirical distribution of $(X_i)_{i \in [n]}$. $X_i \stackrel{\text{iid}}{\sim} \mu^*$, so by SLLN $\pi_n \stackrel{as}{\to} \mu^*$ and since

$$m_n = \frac{\mu + n\pi_n}{\mu(E) + n} \tag{2}$$

(1.) follows.

It remains to show (X_n) is a Polya sequence with parameter μ , i.e.

$$\Pr[A] = \prod_{x} \mu(x)^{[n(x)]} / \mu(E)^{[n]} \tag{3} \quad \{\{\texttt{eq:polya-s} \ \texttt{eq-meas}\}\}$$

where $A = \{X_i = x_i\}_i$, $n(x) = \#\{i : x_i = x\}$, and the rising factorial $a^{[k]} = a(a+1)\cdots(a+k-1)$. Notice

$$\Pr[A] = \mathbb{E}\left[\Pr[A \mid \mu^*]\right] = \mathbb{E}\left[\prod_x \mu^*(x)^{n(x)}\right]$$
(4)

Since μ^* is Ferguson, viewing $E = \bigsqcup_{x \in E} \{x\}$ as a partition we have $(\mu^*(x))_{x \in E} \sim \text{Dir}((\mu(x))_{x \in E})$ so the RHS is the $(n(x))_{x \in E}$ moment of the Dirichlet distribution, which is equal to

$$\mathbb{E}\left[\prod_{x} \mu^{*}(x)^{n(x)}\right] = \frac{\Gamma(\mu(E))}{\Gamma(\mu(E) + n)} \prod_{x} \frac{\Gamma(\mu(x) + n(x))}{\Gamma(\mu(x))} = \frac{1}{\mu(E)^{[n]}} \prod_{x} \mu(x)^{[n(x)]}$$
(5)

as required by Eq. (3).

General E follows from approximation argument.

We leave the discreteness part of (1.) as an exercise, noting that similar to how Dirichlets can be defined as a set of independent Gammas normalized by their sum (Proposition 4 (Constructing Dirichlet from Gammas)) we would expect the Dirichlet process / Ferguson random measures to be definable as a gamma process with independent "increments" divided by their sum.

Exercise 11. Prove every Ferguson random measure is discrete. (Hint: argue using moments).

Remark 12. If (X_n) a Polya sequence, then it is a mixture of iid sequences, and is exchangeable i.e. $(X_i) \stackrel{d}{=} (X_{\sigma(i)})$ (see Eq. (3))

1.2 Invariant measures and Kakutani's Theorem

For a finite group G, the measure $\mu(g) = \frac{1}{\#G}$ is left and right translation invariant i.e. $\mu(gA) = \mu(A) = \mu(Ag)$.

In fact, all compact groups have unique translation invariant measure called the Haar measure. For example $Z_{ij} \stackrel{\text{iid}}{\sim} N(0,1)$ for $i,j \in [n]$ and X the Gram-Schmidt orthonormalization of the rows of Z. Then $XU \stackrel{d}{=} UX$ for all $U \in O(n)$, so X has Haar measure on compact group O(n).

Definition 13

A $topological\ vector\ space\ (TVS)$ is a vector space equipped with a topology such that vector space operations are jointly continuous.

Example 14

 \mathbb{R}^n with standard topology, any Banach space.

Definition 15

{def:equicon | A family &

tinuous}

A family \mathfrak{G} of linear transformations on TVS \mathfrak{X} is equicontinuous on subset $K \subset \mathfrak{X}$ if for every neighborhood V of the origin, there exists a neighborhood U of the origin such that

$$\forall k_1, k_2 \in K : k_1 - k_2 \in U \Rightarrow \mathfrak{G}(k_1 - k_2) \subset V \tag{6}$$

That is, $T(k_1 - k_2) \in V$ for all $T \in \mathfrak{G}$.

Definition 16

A locally convex topological vector space (LCTVS) is a TVS with a local base of absolutely convex absorbing sets at the origin.

To construct Haar measure for any compact group, we will need a fix point theorem due to Kakutani.

Theorem 17 (Kakutani Fix Point Theorem)

 $\{ \texttt{thm:} \texttt{kakutan} \\ \texttt{i} \}$

K compact convex subset of LCTVS \mathfrak{X} , \mathfrak{G} group of linear transforms equicontinuous on K and such that $\mathfrak{G}(K) \subset K$, then $\exists p \in K$ such that

$$\mathfrak{G}(p) = \{p\} \tag{7}$$

Proof. • By Zorn's lemma applied to chains $(K_{\lambda})_{\lambda}$ (note $K_a \subset K_b$ for a < b), \exists minimal $K_1 \subset K$ such that $K_1 \neq \emptyset$ and $\mathfrak{G}(K_1) \subset K_1$.

- If K_1 is a single point, then proof is complete.
- Otherwise, by minimality the compact (because is continuous) set $K_1 K_1$ contains a point other than the origin, so exists $V \in N(0)$ such that $\bar{V} \not\supset K_1 K_1$.
- For some $|\alpha| \leq 1$, there is a convex neighborhood $V_1 \in N(0)$ such that $\alpha V_1 \subset V$.
- By equicontinuity of \mathfrak{G} on $K \supset K_1$, there is $U_1 \in N(0)$ such that for $k_1, k_2 \in K$ and $k_1 k_2 \in U_1$ we have $\mathfrak{G}(k_1 k_2) \subset V_1 >$
- Because $T \in \mathfrak{G}$ is invertible (\mathfrak{G} is a group), T maps open sets to open sets and $T(A \cap B) = TA \cap TB$ for sets A, B.
- Since T is linear, for any A

$$T\operatorname{conv}(A) = \operatorname{conv}(TA)$$
 (8)

• So

$$U_2 := \operatorname{conv}(\mathfrak{G}U_1 \cap (K_1 - K_1)) = \operatorname{conv}(\mathfrak{G}(U_1 \cap (K_1 - K_1))) \subset V$$

is relatively open in $K_1 - K_1$ and satisfies $\mathfrak{G}U_2 = U_2 \not\supset K_1 - K_1$.

- By continuity, $\mathfrak{G}U_2 = \overline{\mathfrak{G}U_2}$.
- Let $\delta = \inf\{a : a > 0, aU_2 \supset K 1 K_1\} \ge 1$, by compactness $\delta < \infty$. Let $U := \delta U_2$.
- For each $\varepsilon \in (0,1)$

$$(1+\varepsilon)U\supset K_1-K_1\not\subset (1-\varepsilon)\overline{U}$$

- Because $(1-1/4n)\bar{U} \not\supset K_1 K_1$, we have $K_2 \neq K 1$.
- K_2 is closed and convex
- Further, since $T(a\bar{U}) \subset a\bar{U}$ for $T \in \mathfrak{G}$, we have

$$T(a\bar{U}+k) \subset a\bar{U}+Tk$$
 for all $T \in \mathfrak{G}, k \in K_1$

• Recalling $TK_1 \subset K_1$ for $T \in \mathfrak{G}$ and that \mathfrak{G} is a group, we find that $TK_1 = K_1\mathfrak{G}K_2 = \underline{\text{a contradic-}}$ Finish tion

2 Lecture 2: Measure theory

2020-01-23

Throughout our discussion, all topological spaces are assumed Hausdorff unless explicitly noted otherwise.

2.1 Construction of Haar measure

Definition 18

A topological group is a group equipped with a topology such that the group operations $(g, h) \mapsto gh$ and $g \mapsto g^{-1}$ are continuous.

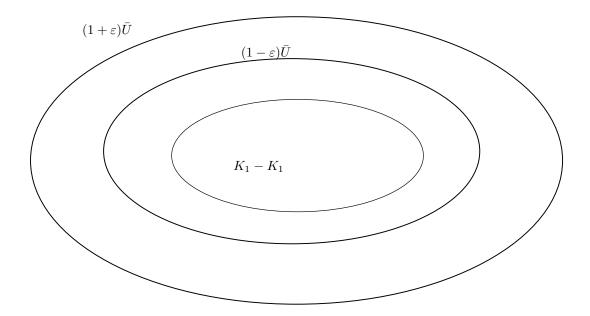


Figure 1: Sketch of proof of Kakutani's theorem

Theorem 19 (Existence of Haar Measure)

{thm:haar-me
asure}

Let G be a compact topological group and $\mathcal{C}(G)$ the set of continuous maps $G \to \mathbb{R}$. Then there is a unique linear form $m : \mathcal{C}(G) \to \mathbb{R}$ such that

- 1. $m(f) \ge 0$ for $f \ge 0$ (positive)
- 2. m(1) = 1 (normalized)
- 3. m(sf) = m(f) where $sf(g) = f(s^{-1}g)$ for $s, g \in G$ (left invariant)
- 4. $m(f_s) = m(f)$ where $f_s(g) = f(gs)$ (right invariant)

m is called the Haar measure on G.

We will need the following theorem to relate compactness with equicontinuity:

Theorem 20 (Generalization of Arzela-Ascoli)

 $\label{thm:arzela-ascoli} $\{ \mbox{thm:arzela-} \\ \mbox{ascoli} $\} $$

Let X be a compact Hausdorff space. A subset of \mathbb{R} -valued continuous functions $F \subset \mathcal{C}(X)$ is relatively compact in topology induced by uniform norm $\|\cdot\|_{\infty} \Leftrightarrow F$ is equicontinuous and pointwise bounded.

Proof of Theorem 19. Fix $f \in \mathcal{C}(G)$ and let \mathcal{C}_f denote the convex hull of all left translates of f, i.e. $g \in \mathcal{C}_f$ are finite sums of form

$$g(x) = \sum_{\text{finite}} a_i f(s_i x), \qquad a_i > 0, \sum_{\text{finite}} a_i = 1, s_i \in G$$

Clearly $||g||_{\infty} \leq ||f||_{\infty} < \infty$, thus $C_f(x) = \{g(x) : g \in C_f\}$ is bounded for all $x \in G$ hence C_f is pointwise bounded.

As f is a continuous function on compact G, it is uniformly continuous hence for $\varepsilon > 0$ there exists a neighborhood V_{ε} of the identity $e \in G$ such that

$$y^{-1}x \in V_{\varepsilon} \Rightarrow |f(x) - f(y)| \le \varepsilon$$

Since $(s^{-1}y)^{-1}s^{-1}x = y^{-1}x$, we also have

$$y^{-1}x \in V_{\varepsilon} \Rightarrow |f(y) - f(x)| < \varepsilon$$

Since $g \in \mathcal{C}_f$ are convex combinations of $_sf$, by the triangle inequality

$$y^{-1}x \in V_{\varepsilon} \Rightarrow |g(y) - g(x)| < \varepsilon$$

{{eq:combo-c lose-to-cons tant}}

As this works for any $g \in \mathcal{C}_f$, we have that \mathcal{C}_f is equicontinuous.

By Theorem 20 (Generalization of Arzela-Ascoli), C_f is relatively compact in C(G), so its closure $K_f := \overline{C_f}$ is compact (and still convex).

Consider G acting on $\mathcal{C}(G)$ by left translation $f \mapsto {}_s f$. Notice $G\mathcal{C}_f \subset \mathcal{C}_f$ (as \mathcal{C}_f already contains all finite convex combinations of all left translations of f) and hence $GK_f \subset K_f$ as well.

Furthermore, $\| {}_s f - {}_s g \|_{\infty} = \| f - g \|_{\infty}$ so G acts as a group of isometries on $\mathcal{C}(G)$. In particular, this group is equicontinuous (with the same U = V in Definition 15).

Taking $\mathfrak{G} = G$ and $K = K_f$ in Theorem 17 (Kakutani Fix Point Theorem), there is a fixed point $g \in K_f$ of this action of G on K_f which satisfies

$$_{s}g=g\;(\forall s\in G)\quad\Rightarrow\quad g(s^{-1})=_{s}g(e)=g(e)=c\;(\forall s\in G)$$

for some constant $c \in \mathbb{R}$ (which we will later use to define $m(f) \coloneqq c$).

We first show there is only one constant function in K_f , so the fix point $Gg = \{g\} = \{c\mathbb{1}\}$ is unique and m(f) = c is well defined. For any constant function $c\mathbb{1} \in K_f$ and $\varepsilon > 0$, we can (because $K_f = \overline{C_f}$) find $\{s_1, \ldots, s_n\} \subset G$ and $a_i > 0$ such that

$$\sum_{i=1}^{n} a_i = 1, \quad \text{and} \quad \left| c - \sum_{i=1}^{n} a_i f(s_i x) \right| < \varepsilon \qquad (\forall x \in G)$$
(9)

for any $\varepsilon > 0$.

Similarly, consider the same construction as before expect now use right translations of f (i.e. using the opposite group G' of G, or the function $f' = f(x^{-1})$, obtaining relatively compact set \mathcal{C}'_f with compact convex closure K'_f with fix point $g' = c' \mathbb{1}$). Approximating $c' \mathbb{1}$ using \mathcal{C}'_f , we have

$$\left| c' - \sum_{j} b_{j} f(xt_{j}) \right| < \varepsilon \qquad \text{(for some } t_{j} \in G, \, b_{j} > 0 \text{ with } \sum_{j} b_{j} = 1) \tag{10}$$

Opposite group

The opposite group g' of the group G is the group that coincides with G as a set but has group operation $(x,y)\mapsto y^{-1}x^{-1}$

Summing over i

$$\left| c' - \sum_{i,j} a_i b_j f(s_i t_j) \right| < \varepsilon \sum_i a_i = \varepsilon$$

Operating symmetrically on Eq. (9) (multiply by b_i and put $x = t_i$) shows

$$\left| c - \sum_{i,j} a_i b_j f(s_i t_j) \right| < \varepsilon$$

Together, we have $|c'-c| < 2\varepsilon$ so taking $\varepsilon \to 0$ shows c'=c. Since $c\mathbbm{1} \in K_f$ was an arbitrary constant function, we have that the constant function in K_f is actually unique and so the function $m(f) := c \in K_f$ is well defined. Moreover, $m(f)\mathbbm{1}$ is the *only* constant function which can be arbitrary well approximated by convex combinations of left or right translates of f.

The following properties are obvious:

- m(1) = 1 since $K_f = \{1\}$ for f = 1
- $m(f) \ge 0$ if $f \ge 0$
- $m(sf) = m(f) = m(f_s)$ (since $K_{sf} = K_f$, $K'_f = K'_{f_s}$, and uniqueness of $m(f)\mathbb{1}$ being the only constant function approximable by both K_f and K'_f)

• m(af) = am(f) for any $a \in \mathbb{R}$ (since $K_{af} = K_f$)

To show m is linear, it suffices (due to the last bullet above) to show that m is additive. Fix $f, g \in \mathcal{C}(G)$. Approximate m(f) using K_f to get

$$m(f) - \sum_{\text{finite}} a_i f(s_i x)$$

$$(11) \quad \{\{\text{eq:approx-mf-using-Kf}\}\}$$

Define $h(x) = \sum_{\text{finite}} a_i g(s_i x)$ using the same a_i and a_i and approximate m(h) using C_h to get

$$\left| m(h) - \sum_{\text{finite}} b_j h(t_j x) \right| < \varepsilon$$

Since $h \in \mathcal{C}_g$, we have $\mathcal{C}_h \subset \mathcal{C}_g$ hence $K_h \subset K_g$. But $m(g)\mathbb{1} \in K_g$ is the only constant function so m(h) = m(g) and (after expanding the definition of h) we have

$$\left| m(g) - \sum_{i,j < \infty} a_i b_j g(s_i t_j x) \right| < \varepsilon$$

On the other hand, multiplying Eq. (11) by b_j replacing x with $t_j x$, summing over j, and finally adding with the above inequality gives

$$|m(f) + m(g) - \sum_{i,j} a_i b_j (f+g)(s_i t_j x)| < 2\varepsilon$$

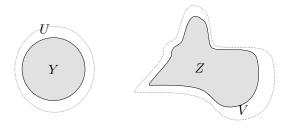
Thus $m(f) + m(g) \in K_{f+g}$, establishing additivity. Note that the only constant in K_{f+g} is $(m(f) + m(g))\mathbb{1}$.

2.2 Facts from topology

We now want to head towards some integration against probability measures defined on spaces more abstract than \mathbb{R}^n .

Definition 21

A topological space X is normal if for any disjoint closed sets Y and Z there exists disjoint open sets U and V such that $Y \subset U$ and $Z \subset V$.



{fig:normaltopologicalspace}

Figure 2: Normal topological spaces admit separating closed sets with two disjoint open sets

Definition 22

X is completely regular (Tychonoff if X is also Hausdorff) if for all $y \in X$ and every closed $Z \subset X \setminus \{y\}$ there exists $f: X \to [0,1]$ continuous such that f(y) = 0 and f(z) = 1 for all $z \in Z$. We say y and Z are separated by a (Urysohn) function.

Corollary 23 (Urysohn's Lemma)

Every normal space is completely regular.

Lemma 24

A compact (Hausdorff) space is normal hence completely regular.

Proof. Fix disjoint closed Y and Z and let $y \in Y$. Consider the open cover of Z given by $\{V_{y,z} : z \in Z\}$ where each $V_{y,z} \in N(z)$ is disjoint from some $U_{y,z} \in N(y)$ (existence ensured by Hausdorff). By compactness, there exists a finite subcover $\{V_{y,z_i}\}_{i=1}^n$. For each of these V_{y,z_i} , let $U_{y,z_i} \in N(y)$ denote the corresponding disjoint neighborhood of y and consider

$$U_y' = \bigcap_{i=1}^n U_{y,z_i} \in N(y)$$

 U'_y is open because it is the intersection of finitely many open sets. It is also disjoint from

$$V_y' \coloneqq \bigcup_{i=1}^n V_{y,z_i}$$

which contains B and is also open.

Now consider the open cover $\{U_y':y\in Y\}$, let $\{U_{y_i}'\}_{i=1}^n$ be a finite subcover, and let $U=\cup_{i=1}^n U_{y_i}'$. Analogously, let $V=\cap_{i=1}^n V_{y_i}'$ where V_y' is given above (open cover of B and disjoint from U_y'). Then $U \supset Y$ and $V \supset Z$ provide two disjoint separating open sets.

Lemma 25

{lem:complet

ely-regularequals-initi

al-topo-cts}

A topological space (X,τ) is completely regular (i.e. Tychonoff) space iff the original topology coincides with the initial topology $\sigma(X,\mathcal{C}(X))$ i.e. the smallest topology that makes every function in $\mathcal{C}(X)$ continuous.

Proof. We only show \Rightarrow . Let U be τ -open and for $x \in U$ pick an Urysohn function $f \in \mathcal{C}(X)$ such that f(x) = 0 and $f(U^c) = 1$. Then $V_x = \{y : f(y) < 1\} = f^{-1}((-\infty, 1))$ is a $\sigma(X, \mathcal{C}(X))$ -open neighborhood of x contained in U, so $U = \bigcup_{x \in U} V_x$ is $\sigma(X, \mathcal{C}(X))$ -open. Since $\sigma(X, \mathcal{C}(X))$ is minimal, we have $\tau = \sigma(X, \mathcal{C}(X))$.

2.3 Radon, Borel, and Baire measures

Definition 26

A non-negative set function $m: 2^X \to [0, +\infty]$ on X is an outer measure on X (or Carathéodory outer measure) if:

- 1. $m(\emptyset) = 0$
- 2. $A \subset B \Rightarrow m(A) \leq m(B)$ (monotone) 3. $m(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} m(A_n)$ for all $A_n \subset X$. (countable subadditivity)

Definition 27

Let $m: 2^X \to [0, +\infty]$ be a non-negative set function satisfying $m(\emptyset) = 0$. A set $A \subset X$ is Carathéodory measurable wrt m (Carathéodory m-measurable) if for any $E \subset X$

$$m(E) = m(E \cap A) + m(E \setminus A)$$

We use \mathfrak{M}_m to denote the class of all Carathéodory m-measurable sets.

Theorem 28 (Carthéodory construction)

1. \mathfrak{M}_m is an algebra, m is additive on \mathfrak{M}_m

2. (Finite additivity) For all sequences of pairwise disjoint $A_i \in \mathfrak{M}_m$ and any $E \subset X$

$$m\left(E \cap \bigcup_{i=1}^{n} A_{i}\right) = \sum_{i=1}^{n} m(E \cap A_{i})$$

$$m\left(E \cap \bigcup_{i=1}^{\infty} A_{i}\right) = \sum_{i=1}^{\infty} m(E \cap A_{i}) + \lim_{n \to \infty} m\left(E \cap \bigcup_{i=n}^{\infty} A_{i}\right)$$

3. If m is an outer measure on X, then \mathfrak{M}_m is a σ -algebra, m is countably additive on \mathfrak{M}_m , and m is complete on \mathfrak{M}_m

Remark 29. The outer measure is constructed such that it satisfies countable additivity on the measurable sets \mathfrak{M}_m .

Example 30

Let \mathfrak{X} be a family of subsets of X such that $\emptyset \in \mathfrak{X}$. Given $\tau : \mathfrak{X} \to [0, +\infty]$ with $\tau(\emptyset) = 0$, set

$$m(A) = \inf \left\{ \sum_{n=1}^{\infty} \tau(X_n) : X_n \in \mathfrak{X}, A \subset \bigcup_{n=1}^{\infty} X_n \right\}$$

where $\mathfrak{M}(A) = \infty$ in the absence of such sets X_n . Then m is an outer measure, denoted τ^* . This is where the "outer" comes from: $\cup_n X_n \supset A$ is an outer approximation to A using (potentially overlapping) sets from \mathfrak{X} hence $\sum_{n=1}^{\infty} \tau(X_n)$ is an overapproximation to the "size" of A. m(A) is the best (i.e. smallest) overapproximation.

Recall the Borel σ -algebra, denoted $\mathcal{B}(X)$, is generated by all open sets.

Definition 31

The Baire σ -algebra, denoted by $\mathcal{B}a(X)$, is generated by sets of the form

where $f \in \mathcal{C}(X)$ (called functionally open sets).

Remark 32. $\mathcal{B}a(X)$ is the smallest σ -algebra where every $f \in \mathcal{C}(X)$ is measurable. It coincides (via a truncation and monotonicity argument) to the smallest one making every $f \in \mathcal{C}_b(X)$ measurable. Contrast this to Lemma 25, which shows that completely regular spaces are those with the smallest topology where every $f \in \mathcal{C}(X)$ is continuous.

Remark 33. Since the functionally open sets can be written as $f^{-1}((0,\infty))$ for continuous f, they are also Borel sets. Therefore, the class of Baire sets are contained in the class of Borel sets.

Lemma 34

In a metric space (X, d), any closed set S is the set of zeros of a continuous function (namely $d_S(x) = \inf_{s \in S} d(x, s)$). Hence, $\mathcal{B}(X) = \mathcal{B}a(X)$.

Lemma 35 (Baire sets are countably determined)

Every $A \in \mathcal{B}a(X)$ is determined by some countable family of functions, i.e. has the form

$$A = \{x : (f_i(x))_{i=1}^{\infty} \in B\}$$
 for some $f_i \in \mathcal{C}(X), B \in \mathcal{B}(\mathbb{R}^{\aleph_0})$

Moreover, every set of this form is Baire and we can take $f_i \in C_b(X)$.

Proof. We first show every set of the same form as A is Baire. True if B is closed, since Lemma 34 allows us to write $B = \phi^{-1}(0)$ for some continuous function $\phi : \mathbb{R}^{\aleph_0} \to \mathbb{R}$ so $\psi = x \mapsto \phi((f_n(x))_{n \ge 1})$ is continuous hence $A = \psi^{-1}(0)$ is also closed. But this is the converse.

For any fixed $\{f_n\}_{n\geq 1}$, the class of sets $B\in\mathcal{B}(\mathbb{R}^{\aleph_0})$ satisfying

$$\{x: (f_i(x))_{i>1} \in B\} \in \mathcal{B}a(X)$$

{lem:metricspace-closed -set-variety is a σ -algebra containing $B = \prod_i (-\infty, a_i)$ where $a_i \neq \infty$ for only finitely many i. This is a basis for $\mathcal{B}(\mathbb{R}^{\aleph_0})$, thus $\mathcal{B}a(X)$ contains it and the two coincide (recall $\mathcal{B}a \subset \mathcal{B}$ since functionally determined sets are \mathcal{B} -open).

On the other hand??? \Box

A consequence of the monotone class theorem ???

Throughout, we consider (signed) measures of bounded variation unless explicitly denoted otherwise.

Definition 36

Let X be a topological space.

- A countably additive measure on $\mathcal{B}(X)$ is called a *Borel measure*
- A countably additive measure on $\mathcal{B}a(X)$ is called a *Baire measure*
- A Borel measure μ on X is called *Radon measure* if every $B \in \mathcal{B}(X)$ can be approximated from the inside by compact sets: for $\varepsilon > 0$ exists $K_{\varepsilon} \subset B$ such that $|\mu|(B \setminus K_{\varepsilon}) < \varepsilon$.

When are two Borel measures equal?

Lemma 37

If two Borel measures coincide on all open sets, thne they coincide on all Borel sets.

Proof. Split $\mu = \mu^+ - \mu^-$ and notice that each of the two components are nonnegative and coincide on open sets. By monotone class theorem, $\mu^+ = \mu^-$.

 μ is Radon iff $|\mu|$ is Radon iff both μ^+ and μ^- are Radon.

Inner and outer approximatino of measures on \mathbb{R}^n :

Theorem 38

 $\mu \geq 0$ on $\mathcal{B}(\mathbb{R}^n)$, then any Borel set $B \subset \mathbb{R}^n$ and any $\varepsilon > 0$ exists U_{ε} open and F_{ε} closed such that $F_{\varepsilon} \subset A \subset U_{\varepsilon}$ and $\mu(U_{\varepsilon} \setminus F_{\varepsilon}) < \varepsilon$.

Proof. Let \mathcal{A} the class of all sets $Ain\mathcal{B}$ such that $F_{\varepsilon} \subset A \subset U_e$ and $\mu(U_{\varepsilon} \setminus F_{\varepsilon}) < \varepsilon$ for some closed set F_{ε} and open set U_{ε} .

Every closed A is in \mathcal{A} , since we can take $F_{\varepsilon} = A$ and U_{ε} some open δ -nbd and consider $\delta \to 0$.

It suffices to show that \mathcal{A} is a σ -algebra, since the closed sets generate \mathcal{B} . \mathcal{A} is closed wrt complements, so it remains to verify closure under countable union.

Let $A_j \in \mathcal{A}$, $\varepsilon > 0$. Then exists closed F_j and open U - j such that $F_j \subset A_j \subset U - j$ and $\mu(U_j \setminus F_j) < \varepsilon 2^{-j}$ for $j \in \mathbb{N}$.

The set $U = U_{j=1}^{\infty} U_j$ is open, and $Z_k = U_{j=1}^k F_j$ is closed.

Observe $Z_k \subset \bigcup_{j=1}^{\infty} A_j \subset U$ and for sufficiently large $k \ \mu(U \setminus Z_k) < \varepsilon$.

Indeed, $\mu(\cup_{j=1}^{\infty}\mu_{j}\setminus F_{j})<2$

□ -(?

finish

Definition 39

Set function $\mu \geq 0$ defined on $\mathcal{A} \subset 2^X$ is *tight* on \mathcal{A} if $\forall \varepsilon > 0$ exists compact $K_{\varepsilon} \subset X$ such that $\mu(A) < \varepsilon$ for all $A \in \mathcal{A}$ that does not meet K_{ε} .

Additive set function μ of bounded variation on an algebra is *tight* if its total variation $|\mu|$ is tight.

A Borel measure is tight iff $\forall \varepsilon > 0$ exists compact K_{ε} such that $|\mu|(X \setminus K_{\varepsilon}) < \varepsilon$ (the "total variation measure").

The second definition is necessary to handle Baire sets.

Definition 40

 μ is regular if $\forall A \in \mathcal{A}, \varepsilon > 0$, $\exists F_{\varepsilon}$ closed such that $F_{\varepsilon} \subset A$, $A \setminus F_{\varepsilon} \in \mathcal{A}$, and

Theorem 27 implies any Borel measure on \mathbb{R}_n is regular, and the same proof works to show any Borel measure on metric space is regular.

Corollary 41 (Baire measures are regular)

{corr:bairemeasure-regu
lar}

Every Baire measure μ on topological space X is regular. Moreover, for every Baire set E and $\varepsilon > 0$, there exists a continuous function f on X such that $f^{-1}(0) \subset E$ and $|\mu|(E \setminus f^1(0)) < \varepsilon$.

3 Lecture 3

Theorem 42 (Extension to Radon measure)

{thm:extendtight-to-rad
on}

Suppose an algebra \mathcal{A} of subsets of topological space X contains a base of the topology. Let μ be a regular additive set function of bounded variation on \mathcal{A} . If μ is tight, then it admits a unique extension to a Radon measure on X.

Proof. V.I. Bogachev, "Measure Theory" Theorem 7.3.2

Tightness is important because it says the whole space is inner approximable by a compact set.

Corollary 43

Let X be a completely regular spaace. Then every tight Baire measure μ on X admits a unique extensino to a Radon measure.

Proof. Every Baire measure is regular by Corollary 41, and since X is completely regular, functionally open sets form a base of the topology. Apply Theorem 42.

This allows us to extend measures on the Baire σ -field to measures on the Borel σ -field.

Definition 44

A vector lattice of functions is a linear space of real functions on a nonempty set Ω such that $\max(f,g) \in \mathcal{F}$ for all $f,g \in \mathcal{F}$.

Notice $\min(f, g) = \max(-f, -g) \in \mathcal{F}$ and $|f| \in \mathcal{F}$.

Theorem 45 (Daniell integration)

Let \mathcal{F} be a vector lattice of functions on a set Ω such that $\mathbb{1} \in \mathcal{F}$. Let L be a linear functional on \mathcal{F} with:

- $L(f) \ge f$ for all $f \ge 0$ (positive)
- L(1) = 1
- $L(f_n) \to 0$ for every $f_n \downarrow 0$

Then there exists a unique probability measure μ on $\mathcal{A} = \Sigma(\mathcal{F})$ generated by \mathcal{F} such that $\mathcal{F} \subset \mathcal{L}^1(U)$ and

$$L(f) = \int_{\Omega} f d\mu, \qquad \forall f \in \mathcal{F}$$

Compare this with Riesz representation theorem

For X a compact space, L linear functional on $\mathcal{C}(X)$ with $L(\mathbbm{1})=1$ and $L(f)\geq 0$ for $f\geq 0$ (positive linear functional), then $L(f)=\int_X f d\mu$ with unique regular Borel probability measure μ on X.

The relation is through Dini's theorem: If $\{f_n\} \subset \mathcal{C}(X)$, X compact, and $f_n(x) \downarrow 0$, then $\lim_{n \to \infty} \sup_{x \in X} f_n(x) = 0$.

Proof. Denote \mathcal{L}^+ the set of all bounded functions f of the form $f(x) = \lim_{n \to \infty} f_n(x)$, where $f_n \in \mathcal{F}$ are nonegative and increasing. $\{f_n\}$ is uniformly bounded, hence $\{L(f_n)\}$ is increasing and bounded by properties of L

Let $L(f) = \lim_{n} L(f_n)$. We show that the extended functional is well-defined, coincides on bounded nonnegative functions in \mathcal{F} with the original functional, and possesses the following properties:

- 1. $L(f) \leq L(g)$ for all $f, g \in \mathcal{L}^+$ with $f \leq g$
- 2. $L(f+g) = L(f) + L(g), L(cf) = cL(f) \text{ for all } c \in [0, +\infty)$
- 3. $\min(f,g) \in \mathcal{L}^+$, $\max(f,g) \in \mathcal{L}^+$, and

$$L(f) + L(g) = L(\min(f, g)) + L(\max(f, g))$$

4. $\lim_{n} f_n \in \mathcal{L}^+$ for every uniformly bounded increasing sequence of functions $f_n \in \mathcal{L}^+$, and $L(\lim_{n} f_n) = \lim_{n} L(f_n)$.

Let $\{f_n\}$ and $\{g_k\}$ be two increasing sequences of nonnegative functions in \mathcal{F} with $\lim_n f_n \leq \lim_k g_k$. By linearity and positivity of L, $\lim_n L(f_n) \leq \lim_n L(g_k)$ proving (1).

From the hypotheses of the theorem if $\psi_m \uparrow \psi \in \mathcal{F}$ are all nonnegative functions then $L(\psi_m) \to L(\psi)$ again through linearity: $L(\psi) - L(\psi_m) = L(\psi - \psi_m) \to L(0) = 0$ proving (2).

To verify (4), suppose $f_{k,n} \uparrow f_n \in \mathcal{L}^+$. Let $g_m = \max_{n \leq m} f_{m,n}$, so $g_m \in \mathcal{F}$ increasing and $f_{m,n} \leq g_m \leq f_m$ for $n \leq m$. Therefore $\lim_m f_m = \lim_m g_m \in \mathcal{L}^+$ and by linearity for $n \leq m$

$$L(g_m) \le L(g_{m+1}), \qquad L(f_{m,n}) \le L(g_m) \le L(f_m)$$

Hence $\lim_{m} L(f_m) = \lim_{m} L(g_m) = L(\lim_{m} g_m) = L(\lim_{m} f_m)$

Figure 3: sketch of proof of 4

{fig:sketchof-proof-of4}

Denote by \mathcal{G} the class of all sets G with $\mathbb{1}_G \in \mathcal{L}^+$, and for $G \in \mathcal{G}$ define $\mu(G) = L(\mathbb{1}_G)$. Observe that min/max convert to indicators, so by (3) the class \mathcal{G} is closed wrt finite intersection/union, and hence by countable union by (4).

Furthermore, μ is nonnegative monotone additive function on \mathcal{G} , with inclusion-exclusion, and $\mu(G_n) \uparrow \mu(G)$.

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According to (TODO: Ref: thm 20) and closure of \mathcal{G} under countable union, the function

$$\mu^*(A) = \inf\{\mu(G) : G \in \mathcal{G}, A \subset G\}$$

is countably additive measure on the class

$$\mathcal{B} = \{ B \subset \Omega : \mu^*(B) + \mu^*(\Omega \setminus B) = 1 \}$$

Uncertain about above theorem

Should check details of section 1.5 Borgachev

For (iii) we verify $\mathcal{A} = \sigma(\mathcal{F}) \subset \mathcal{B}$. If $f \in \mathcal{L}^+$, then $\{f > g\} \in \mathcal{G}$ for all c since

$$\mathbb{1}\{f > c\} = \lim_{n} \min(1, n \max(f - c, 0))$$

Hence $f \in \mathcal{L}^+$ are measurable wrt $\sigma(\mathcal{G})$, but they are also measurable wrt $\sigma(\mathcal{F})$ (since they are monotone limits of things in \mathcal{F}), so $\mathcal{G} \subset \sigma(\mathcal{L}^+) = \sigma(\mathcal{F})$ and by Dynkin π - λ we have $\sigma(\mathcal{G}) = \sigma(\mathcal{F}) = \mathcal{A}$. Thus, it suffices to show $\mathcal{G} \subset \mathcal{B}$.

For $G \in \mathcal{G}$, let $f_n \uparrow \mathbb{1}_G$ so

$$\mu^*(G) = \mu(G) = \lim_{n \to \infty} L(f_n)$$

and since $\mu^*(G) + \mu^*(\Omega \setminus G) \ge 1$, to show $G \in \mathcal{B}$ it suffices to prove $\mu^*(G) + \mu^*(\Omega \setminus G) \le 1$ i.e.

$$\mu^*(\Omega \setminus G) \le \lim_n L(\mathbb{1} - f_n)$$

But since $\mathbb{1} - f_n \downarrow \mathbb{1}_{\Omega \setminus G}$ and $U_c = \{1 - f_n > c\}$ contains $\Omega \setminus G$ hence belongs to \mathcal{G} , therefore

$$\mathbb{1}_{U_c} \le c^{-1}(\mathbb{1} - f_n)$$

$$\mu^*(\Omega \setminus G) \le \mu(U_c) = L(\mathbb{1}_{U_c}) \le c^{-1}L(1 - f_n)$$

Take $c \to 1$ and $n \to \infty$.

It remains to prove $\mathcal{F} \subset \mathcal{L}^1(\mu)$ and that $L(f) = \int_{\Omega} f d\mu$. Approximate $f \in \mathcal{L}^+$ with $f \leq 1$ by simple functions

$$f_n = \sum_{j=1}^{2^n - 1} j 2^{-n} \mathbb{1} \{ j 2^{-n} < x < (j+1)2^{-n} \}$$

$$L(f_n) = \sum_{j=1}^{2^n - 1} j 2^{-n} \mu \{ j 2^{-n} < x < (j+1)2^{-n} \}$$

Hence

$$L(f_n) = \int_{\Omega} f_n d\mu$$

TODO: finish

The uniqueness of μ satisfying ?? follows from the fact that it is uniquely determined on the class \mathcal{G} , which is closed wrt finite intersections and generates \mathcal{A} .

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